

A TARGET TRACKER USING A DOPPLER COMPENSATED
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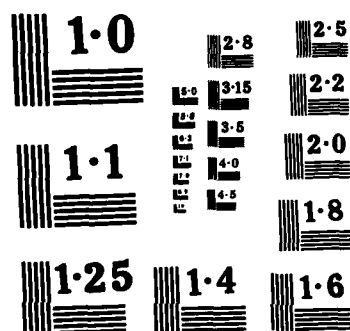
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**RESEARCH REPORT ON A TARGET TRACKER
USING A DOPPLER COMPENSATED
CORRELATION TECHNIQUE**

by

B. R. Eldridge

**Prepared in Response to: Contract N00014-81-C-0535
with
The Office of Naval Research
May 29, 1985**

**Tetra Tech/Honeywell, Inc
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PREFACE

This document is the final research report on the investigation of a mathematical algorithm to do target tracking, using doppler compensated correlation techniques on input time series streams from several passive acoustic sensors. The algorithm was developed and programmed into a testbed on a VAX-750 computer and was tested using simulated time series data generated by the Tetra Tech Broadband Signal Simulator. Algorithm performance proved dissapointing due to: (1) Numerical instabilities induced by structural anomolies in the sample signal autocorrelation function; (2) The extreme sensitivity of objective function to choice of signal characteristics and processing parameters; (3) Computational intensity of the algorithm.

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1. INTRODUCTION

For the past year, Tetra Tech has been involved in the development and analysis of an algorithm for tracking maneuvering submarines using Doppler compensated correlation techniques. The intended goal was the identification of an algorithm which would be more responsive to target kinematic changes than the usual Kalman filter, thereby providing timely and accurate estimates of target position course and speed at various times along the track. The work was carried out under contract N00014-81-C-0535 with the Office of Naval Research.

The algorithm was developed and a testbed computer code was generated for implementing and testing the algorithm. This program was written in FORTRAN-77 on a VAX-750 computer and is set up to use the Tetra Tech Broadband Signal Simulator (BSS) outputs as its input time series. No attempt was made in this effort to use actual at sea data. The FORTRAN listings of the testbed program are included in Appendix A of this report.

The algorithm research carried out by Tetra Tech assumed the availability of base banded, bandlimited digitally sampled time series from several passive acoustic sensors. The availability of information such as bearings and time delays were not assumed in the testbed program, since it was felt that working directly with the time series data would provide a more convenient means of implementing a maneuvering target tracker. Due to the general structure of the algorithm, additional measurement types such as those mentioned above can be easily added if desired.

As is well known, one of the characteristics of a sequential or Kalman type of estimating scheme is the tendency of the estimator to build up "inertia" and thereby make it unresponsive to changes in target course and speed after long periods of tracking time. This may be overcome by certain ad hoc schemes such as frequent reinitializations or possibly by tampering with the weights so as to cause the algorithm to have a shorter memory, etc.

In view of this it seemed reasonable to employ an estimation scheme which works directly with selected blocks of time series data in a batch mode, and which can be highly overlapped from estimation to estimation. Once the algorithm has been initialized and is operation, the previous estimate of target state can be used to initialize the trial solution for the current estimation. The covariance matrix of the initialization state is not carried over and therefore the process is without a memory. However, if sufficient overlap in the input time series is used, the output states should show minimal change from estimation to estimation while still reflecting the most current information available from the time series.

Time series generated by a single moving source and received at two or more spacially separated sensors will exhibit different Doppler and time delay characteristics at each of the receivers. These characteristics are, of course, dependent on sensor-target geometry and kinematics, as well as sound propagation physics. By applying the appropriate time and Doppler compensation to the received time series, pairwise time series correlations between sensors can be maximized. By linking the time and Doppler compensation to assumed target motion, one can adjust the target state parameters to maximize (or minimize) an appropriately chosen function of the corresponding pairwise cross correlation estimates.

The assumption of digitally sampled time series sampled at a uniform sample rate suggests that the time and Doppler compensation be done in the time domain using an interpolative resampling technique. This involves estimating time series values whose sample times lie between the discrete sample times of the input time series. For band limited signals, the well known Sampling Theorem provides a rational means of performing the required interpolation using neighboring time series points and the sinc function as an interpolating function. This, in effect, generates a piecewise continuous representation of the original time series thereby allowing resampling at arbitrary times which are in concert

with trial target state parameters. Also, such a scheme allows analytic evaluation of the gradient vector with respect to the state variables and weighting vectors.

As has been mentioned above, Tetra tech has implemented these ideas into a testbed algorithm on the VAX-750 digital computer. Inputs to the algorithm consist of up to 10 channels of time series data. For each channel, the algorithm requires an estimation of signal to noise ratio (SNR) along with the standard deviation of the SNR for that channel. The algorithm is also sensitive to such inputs as integration time, processing bandwidth and center frequency, station location, sound speed in water, time series overlap, and initialization parameters such as position course and speed. The target kinematic model consists of polynomials in water time of up to degree 5 for x,y and z. The order of the polynomials is user specified. Model output consists of estimated target parameters, their associated error variances, and the size and orientation of the 2- σ containment ellipsoid.

The algorithm has been tested using simulated time series generated by the BSS. The BSS can emulate the complex time series generated by a moving target having user specified kinematic and spectral output signal characteristics.

Section 2 of this report gives an overall description of the workings of the algorithm. Section 3 describes the Gauss-Newton estimation scheme as it applies to this effort, and Section 4 outlines and justifies the doppler compensation scheme that we have employed. Section 5 contains the objective function minimization process description and algorithm performance is reported in Section 6. Finally, Section 7 presents the summary and conclusions of this effort. Appendix A contains the FORTRAN listings of the testbed program developed as part of this effort.

2. ALGORITHM DESCRIPTION

The algorithm estimation scheme assumes the availability of several channels of bandlimited, basebanded, discretely sampled digital data for which all of the pertinent parameters such as sample rate, bandwidth, and center frequency are known, and all of which contain signal from a common emitter. For the purposes of this analysis, we will assume that the noise on each channel is mutually uncorrelated. We will also assume that the target is moving along some 3-dimensional trajectory, which is given by the vector function $P(s;t)$, where s is the state vector to be estimated and t is the time along the trajectory.

The testbed version of the algorithm assumes that each of the components of $P(s;t)$ is an n 'th order polynomial in t , and the state vector s consists of the coefficients of these polynomials. The testbed user may specify n to be any non negative integer up to and including 5. In practice, n is usually chosen to be 1, resulting in linear target motion at constant speed. In this case, $P(s;t)$ is given by

$$P(s;t)=P_0+Vt \quad (2.1)$$

and the state vector s may be represented in transposed form by

$$s=[P_0^T, V^T]^T \quad (2.2)$$

The superscript T denotes the usual matrix transpose operator.

In order to keep the algorithm tractible, we have assumed linear, constant speed sound propagation. More complicated propagation models could have been incorporated, but it was deemed an unnecessary complication at these early stages of algorithm development and feasibility analysis.

The central idea of the algorithm is to minimize a quadratic form of "system functions". The system functions are dependent on the collection of pairwise normalized sample correlation envelope

functions which have been adjusted to account for assumed target kinematics. Each sample correlation envelope function is obtained by correlating the samples from a selected reference channel with modified sets of samples that have been interpolated and resampled from each the other channels comprising the tracking system. The resampling times are calculated as a function of the current value of the state vector, sensor kinematics, the assumed propagation model, and channel signal processing parameters such as center frequency, sample rate, and bandwidth.

To be specific, let us consider a pair of channels, say channels X and Y. Pick a set of samples $\{X_1, X_2, \dots, X_n\}$ from channel X. These samples correspond to arrival times $\{u_1, u_2, \dots, u_n\}$ on channel X. In order to do the motion compensation, we need to calculate the corresponding arrival times $\{v_1, v_2, \dots, v_n\}$ on channel Y. This is done by using the candidate source trajectory $P(s; t)$ to calculate emitter times $\{t_1, t_2, \dots, t_n\}$ corresponding to the X channel arrival times $\{u_1, u_2, \dots, u_n\}$, and using these emitter times to project the corresponding arrival times $\{v_1, v_2, \dots, v_n\}$ on channel Y. We then interpolate and resample the Y channel at the $\{v_1, v_2, \dots, v_n\}$ thereby obtaining a new set of samples $\{Y_1, Y_2, \dots, Y_n\}$. The interpolation is accomplished using a truncated sinc function as an interpolating function. Details of the interpolation scheme are provided in Section 4 of this report.

The magnitude squared cross correlation estimate δ_{xy} is then calculated by

$$\delta_{xy} = |\sum X_i Y_i^*| / \{\sum |X_i|^2 \sum |Y_i|^2\} \quad (2.3)$$

where the three sums in the above expression are taken over $i=1, 2, \dots, n$ and the superscript (*) denotes complex conjugation. The values of the δ_{xy} so obtained are used to form the aforementioned system functions F_{xy} which are used in the minimization process. The system functions are given by

$$F_{xy} = \ln(G_{xy} / \delta_{xy}) \quad (2.4)$$

where G_{xy} is the a priori expected value of δ_{xy} . G_{xy} is related to the input SNR estimates on each channel by

$$G_{xy} = [(1 + \text{SNR}_x^{-1})(1 + \text{SNR}_y^{-1})]^{-1} \quad (2.6)$$

Note that the F_{xy} are chosen such that they have value 0 when given error free information.

Finally, $Q(\mathbf{s})$, a positive definite quadratic form of the F_{xy} , is formed over all channel pairs, and by using gradient methods, the state vector \mathbf{s} is adjusted so as to minimize $Q(\mathbf{s})$. The vector \mathbf{s}_0 which minimizes $Q(\mathbf{s})$ is taken as the state estimate.

A fallout of the minimization process is an estimate of the state covariance matrix. This matrix is used to calculate the ellipsoidal containment region which provides the user with a geometric indication of algorithm performance.

3. THE ESTIMATION SCHEME

3.1 Preliminary Details and Notation

this section details the estimation scheme in rather general mathematical terms. Preliminary to the discussion we establish the following notation which will be used throughout the remainder of this report.

If \mathbf{X} and \mathbf{Y} are n -dimensional complex valued vectors, the complex inner product of \mathbf{X} and \mathbf{Y} , denoted by $\langle \mathbf{X}, \mathbf{Y} \rangle$, is defined by

$$\langle \mathbf{X}, \mathbf{Y} \rangle = \sum X_i Y_i^* \quad (3.1)$$

where the sum is taken over $i=1,2,\dots,n$, and the X_i and Y_i are the complex valued components of \mathbf{X} and \mathbf{Y} , respectively. The $(*)$ notation denotes complex conjugation. Note that the complex inner product is conjugate symmetric in that the following relationship holds:

$$\langle \mathbf{Y}, \mathbf{X} \rangle = \langle \mathbf{X}, \mathbf{Y} \rangle^* \quad (3.2)$$

We define the norm of \mathbf{X} , denoted by $\|\mathbf{X}\|$, by

$$\|\mathbf{X}\| = \sqrt{\langle \mathbf{X}, \mathbf{X} \rangle} \quad (3.3)$$

In this notation, Equation 2.3 of Section 2 becomes

$$r_{xy} = |\langle \mathbf{X}, \mathbf{Y} \rangle|^2 / (\|\mathbf{X}\|^2 \|\mathbf{Y}\|^2) \quad (3.4)$$

If each of the components of the complex valued vector is a differentiable function of some real parameter θ , then denote the vector consisting of the corresponding derivatives (partial derivatives) by $d\mathbf{X}/d\theta$ ($\partial\mathbf{X}/\partial\theta$). It is easy to verify that the following useful relationship is true

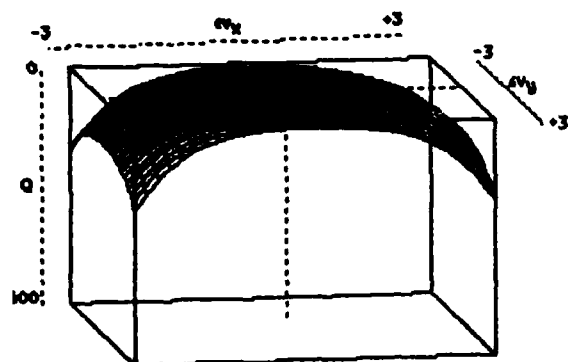


Figure 6.2
Objective Function vs errors in Velocity Probes
 $f_c=20$ Hz
 $T_{int}=20$ sec
 $BW=6$ Hz
Position Offset $= (0,0)$

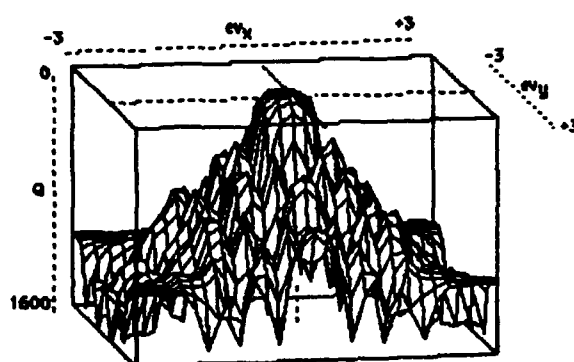


Figure 6.3
Objective Function vs errors in Velocity Probes
 $f_c=20$ Hz
 $T_{int}=100$ sec
 $BW=6$ Hz
Position Offset $= (0,0)$

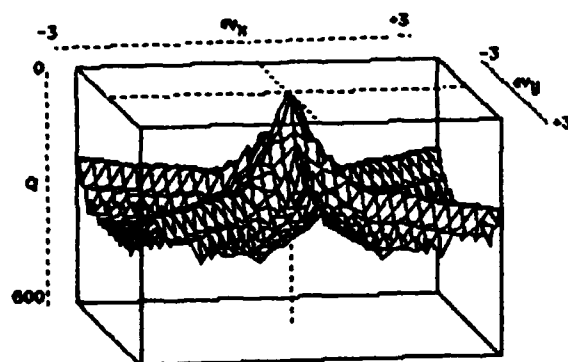


Figure 6.4
Objective Function vs errors in Velocity Probes
 $f_c=230$ Hz
 $T_{int}=20$ sec
 $BW=6$ Hz
Position Offset $= (0,0)$

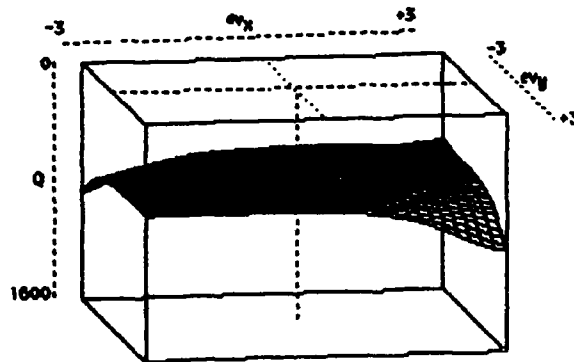


Figure 6.5
Objective Function vs errors in Velocity Probes
 $f_c=20$ Hz
 $T_{int}=20$ sec
 $BW=6$ Hz
Position Offset $= (300,300)$

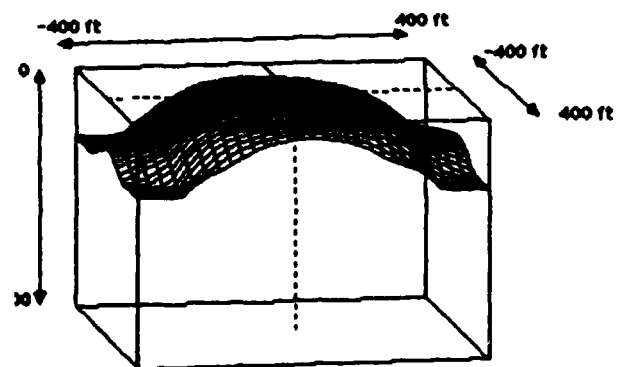


Figure 6.6
Objective Function vs Position Error
 $f_c=20$ Hz
 $T_{int}=20$ sec
 $BW=6$ Hz
Velocity Errors $= (0,0)$

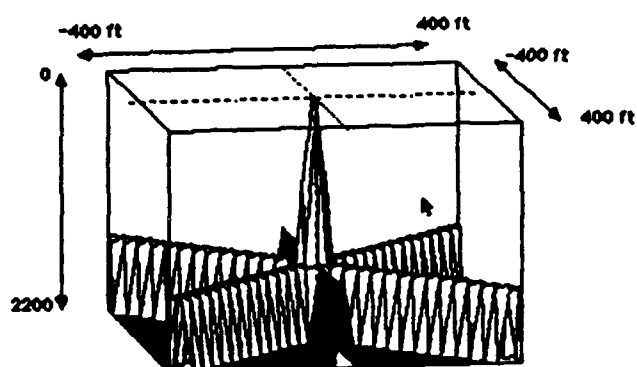


Figure 6.7
Objective Function vs Position Error
 $f_c=80$ Hz
 $T_{int}=20$ sec
 $BW=100$ Hz
Velocity Errors $= (0,0)$

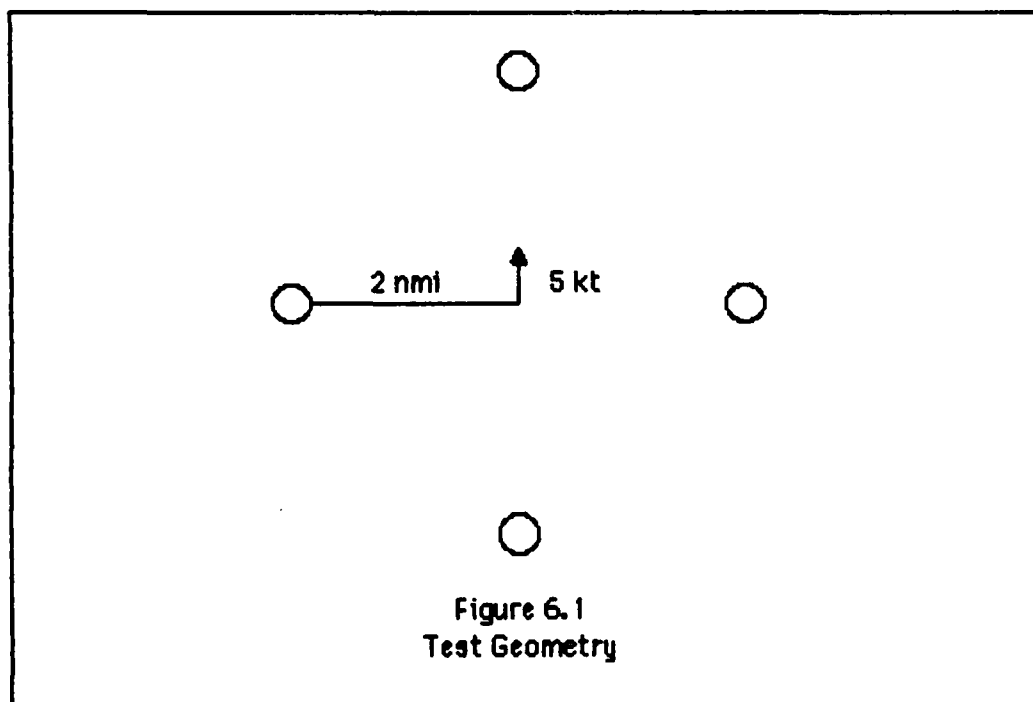
The following figures show the kinds of difficulties one faces even under the best of circumstances. Figures 6.2-6.5 show plots of the objective function versus errors in the probe or trial solution velocities(ft/sec) in both the x and y direction for several sets of processing parameters. The assumed position error is assumed to be fixed at 0 ft. The velocity errors range from ± 3 ft/sec in both the x and y direction. The center of the portion of the x-y plane shown represents 0 error. The z axis represents the values of the objective function and has been inverted for the sake of this presentation. The z axis ranges from 0 (top) to 1600 (bottom). The signal bandwidth is assumed to be 8 Hz.

Figure 6.2 was generated assuming a center frequency of 20 Hz with an integration time of 20 seconds and presents a very clean objective function over the plot range. In Figure 6.3, the integration time has been increased 100 seconds. The resulting plot exhibits a much more spiked peak around (0,0) and the outskirts show a good deal of ripple. Figure 6.4 is similar to Figure 6.2 except that the center frequency of the signal has been shifted to 250 Hz. This band shift has caused the peak to become very sharp with even more ripple evidenced in the outskirts. The processing situations depicted in Figures 6.3 and 6.4 would present problems to the tracker. Figure 6.5 has identical parameters as Figure 6.2 with the exception that the assumed position error has been offset to 300 ft. in both the x and y directions resulting in a much more planar shape and having increased the overall magnitude of the objective function considerably.

Figures 6.6 and 6.7 show plots of the objective function versus errors in the trial position estimates with the geometry in Figure 6.1 applying. The position errors range from ± 400 ft in both the the x and y directions. Figure 6.6 corresponds to a signal bandwidth of 8 Hz and presents a rather smooth function with essentially no unusual structure. In Figure 6.7 we have increased the signal bandwidth to 100 Hz and moved the center frequency to 80 Hz. The resulting plot contains considerable structure with the

parameters, we developed a computer program to generate the expected value of $Q(s)$ for a particular signal autocorrelation function. The outputs are sensitive to the aforementioned signal and processing parameters as well as target/sensor geometry and kinematics. The autocorrelation function we have chosen is triangular on the interval $[-T, T]$ and has a spectral density function of the form $\text{sinc}^2(\omega T/2)$. Its bandwidth is approximately $1/T$. This particular autocorrelation function has the advantage of being integrable in closed form. The program is setup to generate surfaces of expected values of the tracker objective function, holding the position probes fixed and letting the velocity probes vary, or holding the velocity probes fixed and letting the position probes vary.

Some results are presented for the geometry shown in Figure 6.1. Here the target is assumed to be in the center of a square box with the sensors located at the vertices. The distance from the target to each of the sensors is assumed to be 2 nmi. A SNR of 6 dB with a $\sigma_{\text{SNR}}=3$ dB is assumed.



6. ALGORITHM PERFORMANCE

6.1 Parameter Selection

Since the objective function $Q(s)$ is a very complicated function (most of which is carried out in the complex domain) of a number of processing parameters, some thought had to be given to their effects on algorithm behavior. The user has control over such things as processing bandwidth, center frequency, and integration time.

For example, if one chose to process a signal having a sufficiently high center frequency for a sufficiently long period of time, $Q(s)$ could become quite sensitive to trial solution errors in velocity. In fact one would expect to see spike like behaviour in the objective function in the neighborhood of the true velocity components, which could conceivably cause convergence difficulties.

Similarly, a wide bandwidth signal could cause similar spike like behavior in trial solution position estimates. Small positional errors wind up in the ripple of the objective function which wreaks havoc on the convergence process we have chosen.

On the other hand, if the processing band is too narrow, the lack of time delay resolution may not provide any meaningful information to the algorithm, again resulting in poor behavior.

These were in fact some of the problems that were encountered during the early stages of algorithm testing. Since the testbed computer code was newly developed, it was not known whether poor early algorithm performance was due to bugs in the program, or whether the algorithm just did not work, or whether we had just chosen bogus processing parameters. After considerable reexamination, rechecking, and rederiving the mathematics, we decided that they were correct. We could find no bugs in the program and so our only recourse was to give careful scrutiny to our choice of test parameters.

In order to gain insight to the effects of processing

which yields

$$\partial s = -(F_s^T W F_s)^{-1} (F_s^T W F) = C_{ss} (F_s^T W F) \quad (5.5)$$

The advantage of this approach is that it does not require the computation of second derivatives, and that an estimate of the state covariance matrix falls out of the process.

In summary, if s is the current trial solution for the process, we form the next iterate s' by calculating ∂s as in the preceding equation and forming s' by

$$s' = s + \partial s \quad (5.6)$$

To stop the process we check to see if the magnitude of ∂s is within some predescribed tolerance. If so the process is stopped and the current value of the state vector is returned as the solution. If not the process continues until a solution is returned or the maximum iteration count is exceeded.

The entire process is described in Figure 5.1.

5. THE MINIMIZATION PROCESS

5.1 The Iteration Scheme

The function $Q(s)$ is a complicated function of the state vector and measurement vector, the minimization of which does not seem amenable to closed form solutions. We therefore must rely on iterative techniques to solve the problem. The following paragraphs describe the technique we have chosen to accomplish the minimization.

Recall that $Q(s)$ is a quadratic form in the system functions and may be written

$$Q(s) = F^T W F \quad (5.1)$$

where the weighting matrix W is chosen to be the inverse of the covariance matrix of the system residual vector. Under the assumption of slowly varying weights, the gradient vector of $Q(s)$ may be written

$$\nabla Q(s) = 2F_s^T W F. \quad (5.2)$$

At a local minimum we have the necessary condition that

$$\nabla Q(s) = 2F_s^T W F = 0 \quad (5.3)$$

An approach which has been used successfully is demonstrated in the following discussion. Suppose that the algorithm has reached a stage such that s is the current value of the trial state vector. We would like to find the perturbation ∂s to add to s which will improve the estimate. A reasonable approach is to solve the the following perturbed gradient equation for ∂s

$$F_s^T W (F + F_s \partial s) = 0 \quad (5.4)$$

receiver times using $Q_h(t)$ as the interpolating function, thereby generating a set of Y channel samples which reflect the Doppler corrections implied by the current value of the state vector.

In order to obtain the receiver times on the Y channel, we solve the following pair of equations for v_k given u_k :

$$\begin{aligned} u_k &= T_k + |P(s, T_k) - P_x|/c \\ v_k &= T_k + |P(s, T_k) - P_y|/c \end{aligned} \quad (4.6)$$

where P_x and P_y are the respective position vectors of the X and Y channel receivers, c is the speed of sound in water, and T_k is the emitter time. This is done by solving the first equation for T_k using a Newton-Raphson technique, and then using the second equation with the value of T_k so obtained to obtain v_k .

Recall that we are assuming that all of the input time series data has been basebanded from some center frequency f_c . Because of this, a phase correction prior to correlation is required on both channels. This merely amounts to heterodyning the samples back up to their original center frequency f_c . Form the complex vectors X and Y whose k 'th components are given by $\exp(2\pi j f_c (u_k - u_0)) X_{n+k-1}$ and $\exp(2\pi j f_c (v_k - v_0)) Y_{n+k-1}$, respectively. Then δ_{xy} is given by

$$\delta_{xy} = | \langle X, Y \rangle |^2 / (\|X\|^2 \|Y\|^2) \quad (4.7)$$

4.3 System Function Derivatives

The algorithm uses gradient methods to minimize $Q(s)$, which is the quadratic form of the system functions F_{xy} as discussed above. This necessitates the calculation of the derivatives of $Q(s)$ with respect to each of the state variables. Most of the work is done in computing the derivatives of the δ_{xy} with respect to the measurement vector and each of the state variables. The mathematical development of these derivatives is straightforward but extremely tedious and will not be included here.

reconstructed from its samples, provided that the digital sample rate is at least as great as the bandwidth of the signal. The reconstruction uses the "sine x over x" or sinc as an interpolating function. Specifically, if $Z(t)$ is a time series having non zero frequency content only in the interval $[-b/2, b/2]$, and if $Z(t)$ is uniformly sampled over all time at a sample rate $f_d \geq b$, producing samples Z_n , $-\infty \leq n \leq \infty$, and such that Z_0 corresponds to a sample time of 0, then $Z(t)$ is reconstructed exactly from its samples by

$$Z(t) = \sum \text{sinc}\{\pi(f_d t - n)\} Z_n \quad (4.3)$$

The above sum is taken over all n . This, however, involves summing over an infinite number of elements. It therefore seems reasonable to approximate the reconstruction of $Z(t)$ from its samples by using a time limited version of the sinc function. If we define the interpolating function $Q_h(t)$ by

$$Q_h(t) = \begin{cases} \text{sinc}(t), & |t| \leq h \\ 0 & |t| > h \end{cases} \quad (4.4)$$

, then $Z(t)$ may be approximated by

$$Z(t) \approx \sum Q_h\{\pi(f_d t - n)\} Z_n \quad (4.5)$$

which, for any given value of n involves only finite sums.

Let us assume we are working with channels X and Y and we wish to calculate δ_{xy} for the current value of the state vector. Pick a set of reference samples $\{X_n, X_{n+1}, \dots, X_{n+M-1}\}$ from channel X . These samples correspond to the set of channel X receiver times $\{u_n, u_{n+1}, \dots, u_{n+M-1}\}$. We then use $P(s; t)$ and the assumed propagation model to determine the corresponding set of channel Y receiver times $\{v_n, v_{n+1}, \dots, v_{n+M-1}\}$. Channel Y is then interpolated at these

4. DOPPLER COMPENSATION SCHEME

4.1 Time Series Assumptions

The algorithm assumes the availability of M channels of data, and that the time series for each data channel contains basebanded, bandlimited data that has been sampled at at least the Nyquist sampling rate. For the sake of notational simplicity, we will assume that all of the channels have the same center frequency, f_c , and the same digital sampling rate f_d . We also assume that the n 'th sample on each channel occurs at the same time. The time between samples, Δt , is given by

$$\Delta t = 1/f_d \quad (4.1)$$

Therefore if the 0'th sample corresponds to t_0 , then the n 'th sample corresponds to t_n , where

$$t_n = t_0 + n\Delta t \quad (4.2)$$

The algorithm requires an estimate of the signal to noise ratio and its corresponding error variance on each channel of input data. It is assumed that these are independently specified and will be denoted SNR_m , where the subscript m refers to the particular channel designator.

4.2 Resampling and Phase Compensation

In order to effect the proper Doppler compensation, it is necessary to interpolate between samples in the time domain, with the interpolation times reflecting the current value of the state vector.

The well known sampling theorem from signal processing states that a complex valued bandlimited signal can be completely

$$B = (F_S^T W F) \quad (3.18)$$

Suppose we want to add a new measurement set to the algorithm that is independent of those already incorporated and can be described with a single system equation. This merely requires the specification of the corresponding system function and its associated partial derivatives. In keeping with our earlier discussion, let us further assume that the partial of the new system equation with respect to its measurement vector are independent of the state vector. In this case we first form the new scalar weight w by

$$w = 1 / \sum [(\partial F / \partial m_i) \sigma_{m_i}]^2 \quad (3.19)$$

where F is the new system function, the m_i are its associated measurements and

$$\partial F / \partial m_F = [\partial F / \partial m_i]. \quad (3.20)$$

The updated B vector and Z matrix are given by

$$B = B_{old} + w F (\partial F / \partial S) \quad (3.21)$$

$$Z = Z_{old} + w (\partial F / \partial S) (\partial F / \partial S)^T \quad (3.22)$$

where $\partial F / \partial S$ is the vector of partials of F with respect to the elements of the state vector.

$$F_s W F = 0 \quad (3.13)$$

where F_s denotes the matrix whose mn 'th element is $\partial F_m / \partial s_n$. If the above equation holds, a perturbation in the measurement vector ∂m induces a perturbation in the state vector ∂s , which to within first order terms, obeys the relationship

$$F_s^T W (F + F_s \partial s + F_m \partial m) = 0. \quad (3.14)$$

This implies

$$\partial s \approx -(F_s^T W F_s)^{-1} F_s^T W F_m \partial m, \quad (3.15)$$

which yields, after substituting C_{mm}^{-1} for W

$$C_{ss} \approx E(\partial s \partial s^T) = (F_s^T W F_s)^{-1} \quad (3.16)$$

The expression for C_{ss} given above is extremely convenient and can be used to provide geometrical insight to algorithm behaviour. In particular, it is used to derive the ellipsoidal containment region of the current estimate of the state vector.

3.3 Incorporation of Additional Measurements

The form of the estimator used in this algorithm has the advantage that it is easy to incorporate new types of measurements should the need arise. If, for example, the system can provide independent estimates of position and velocity or if another independent sensor comes on line, the new measurements so provided may be entered as an additive partitions to the system weighting matrix and gradient vector. Let us first establish the following notation. Let

$$Z = (F_s^T W F_s) \quad (3.17)$$

We now provide our rationale for our choice of W . Let m denote the vector of measurements associated with F . In our case, each component of m is an SNR estimate from one of the output channels. For any fixed value of s , the perturbation in the vector F induced by a perturbation in the vector m is approximated by

$$\partial F \approx [F_m] \partial m \quad (3.9)$$

Here $[F_m]$ is the matrix whose ij 'th element is $\partial F_i / \partial m_j$. Assuming the error distribution is zero mean with covariance matrix C_{mm} , and that the approximation to ∂F holds over the probable values of m , we may approximate the covariance matrix of F by

$$C_{FF} \approx F_m C_{mm} F_m^T \quad (3.10)$$

The weighting matrix W referred to in Equation (3.8) is chosen to be $(C_{FF})^{-1}$. Therefore the scalar function $Q(s)$ is given by

$$Q(s) = F^T (C_{FF})^{-1} F \quad (3.11)$$

This choice of W has intuitive appeal in that measurements with higher variance get less weight and, therefore, have less of an effect on the final outcome.

We now turn our attention to the approximation of the output state covariance matrix, which will be denoted by C_{ss} . The components of the state vector s_0 that minimizes $Q(s)$ satisfy the equation

$$\partial Q / \partial s_i = 2(\partial F / \partial s_i) W F + F^T (\partial W / \partial s_i) F = 0, i=1,2,\dots,k \quad (3.12)$$

If we assume slowly varying weights which enables us to ignore the second term in the above equation, then s_0 satisfies the system of equations

$$\partial \langle X, Y \rangle / \partial \theta = \langle \partial X / \partial \theta, Y \rangle + \langle X, \partial Y / \partial \theta \rangle. \quad (3.5)$$

The above relationship is indispensable in computing the gradient derivatives during the minimization process.

3.2 Theoretical Development

The general theoretical basis for the estimation scheme is an extension of the techniques of linear optimal estimation theory to the nonlinear case. We evaluate a set of system equations F_{xy} as given by Equation () and then minimize a positive definite quadratic form of the F_{xy} . Suppose we have k such system functions for a particular application. Changing notation for convenience, let the system functions be denoted by F_1, F_2, \dots, F_k and let

$$F(s) = [F_1, F_2, \dots, F_k]^T \quad (3.6)$$

denote the k -dimensional vector of system equations. Note that the solvability of the above equations implies that $n_s \leq k$, where n_s is the dimensionality of the state vector.

If the geometric and signal assumptions are perfectly compatible with the measurement data, there exists a value of the state vector s_0 for which

$$F(s) = 0. \quad (3.7)$$

In general, however the data does not support perfect compatibility, and for each value of the state vector s , $F(s)$ may be considered a vector of F -residuals. An optimal estimate of s is a vector which minimizes a particular quadratic form in the F -residuals. If W is an appropriately chosen positive definite symmetric matrix, the optimal estimate is the value which minimizes the scalar function

$$Q(s) = F^T W F. \quad (3.8)$$

crisscrosses being corresponding to time delays among the various sensors. Again, finding the minimum of such a function if the solution starts off of the main peak presents quite a formidable task to the tracker.

We have presented these results to demonstrate difficulties in choosing a set of operating parameters for which we may have some hope of achieving positive results. To this end and after having a large number of cases as above, we chose to work with a time series centered at 20 Hz containing an 8 Hz signal in 10 Hz wide total processing bin. The signals were generated by the BBS for sensors having the geometry described in Figure 6.1 above, and had an overall SNR of +6dB in the processing band to each of the sensors. Using an integration time of 20 seconds and initializing the algorithm with "truth" at various points along the target track, the algorithm evidenced unstable convergence in nearly every case. Since we had gone over the computer program and the mathematics very carefully and could find no errors, we had to look elsewhere for an explanation of the poor algorithm performance. We believe the answer lies in the form of the system functions.

The Z matrix and B vector defined in Section 3 of this report are used extensively in the Gauss-Newton minimization technique that we chose to implement for this algorithm. Note that both Z and B contain the partial derivatives of the system functions with respect to each of the elements of the state vector. When the system is near a solution, the partials of the system functions are near 0, since they are essentially the derivatives of the signal autocorrelation function at $t=0$. This causes the Z matrix to become numerically unstable as the process nears a solution, thereby causing unpredictable algorithm behaviour. Efforts to accomodate alternate forms of the system functions turned proved unsuccessful because we could not find a tractible method of comparing time series that did not involve correlations. The only other alternative to successfully implement the algorithm would be to incorporate a more sophisticated minimization technique which would overcome the

instability problems. This, too, could have drawbacks since most such algorithms involve several one dimensional searches which require several objective function evaluations which is quite computationally intensive.

We actually tried a method which involves halving the length of the search vector ∂s until $Q_{k+1} < Q_k$ at which time we would recalculate a new search vector and continue the process. This proved to be too computationally intensive even for TW products on the order of 200 and including 4 sensor geometry. We did, however obtain some success in achieving algorithm convergence. However, each evaluation of $Q(s)$ took about 2 minutes CPU time on the VAX, resulting in enormous total algorithm processing times. This was deemed unsatisfactory from the point of view of any practical application.

7. SUMMARY AND CONCLUSIONS

An algorithm to use doppler compensated resampling and correlation techniques on digitally sampled time series was developed and tested using synthetic time series data generated by the Tetra Tech Broadband Signal Simulator (BBS). The algorithm was designed to provide timely track estimates by using highly overlapped time series segments from the receiving sensors in a batch mode, thereby giving the process a short memory and increased responsiveness to target maneuvers. The target motion model consisted of polynomial functions of time of arbitrary order up to 5.

Algorithm performance proved dissapointing for several reasons:

(1) Computational intensity was more than was originally envisioned.

(2) Structure of the correlation functions induced numerical instabilities in the convergence process which could not be easily overcome.

(3) The structure of the objective function is, in general, quite irregular, thereby requiring careful, and perhaps limited, choice of processing parameters in order to have any hope of successful performance.

Some improvement in running time could be achieved by simplifying the resampling technique to use first and perhaps second order time series stretches based on the current value of the position and velocity estimates.

Curing the numerical instabilities seems to be the most difficult hurdle to overcome due to the often unusual structures evidenced in correlation functions. For this reason, we feel that any further endeavors to improve upon such an algorithm would prove to be dissapointing and recommend that further research be discontinued.

APPENDIX A

TESTBED COMPUTER PROGRAM LISTINGS

28-May-1985 17:32:45
14-May-1985 12:14:28

```

0001 PROGRAM TDRIV
0002 INCLUDE 'TRCNTRL.CMN'
0003 COMMON/TRCNTRL/NDEG,NSAMP,TSTRI_REF,TEND_REF,NSKIP,MAXII,EPSILON
0004 1 ,NCHAN,TRUNC,NSTATE,C
0005 INCLUDE 'TRWORK.CMN'
0006 1 DOUBLE PRECISION ZZ,DX,F,FX,BB,RHOSQ
0007 1 COMMON/TRWORK/SV(3,0:10),F,FX(31),BB(31),RHO,RHOSQ,ZZ(961),ITCOUNT,TS
0008 1 ,DX(31),X(31),COV(4,4),DETCOV
0009 INCLUDE 'TRTMS.CMN'
0010 1 COMMON/TRTMS/FILE_REF,BSZ_REF,CFQ_REF,SR_REF,T_REF,POS_REF(3)
0011 1 ,FNS_REF,SNR_REF,SIGSNR_REF
0012 1 ,FILE_RES(32,5),CFQ_RES(5),BSZ_RES(5),SR_RES(5),T_RES(5)
0013 1 ,FOS_RES(3,5),FNS_RES(5),SNR_RES(5),SIGSNR_RES(5),COH(5),BIAS(5)
0014 1 ,SIGRQCOH(5),ZRESBUF(0:4200),T_BUF
0015 1 BYTE FILE_REF,FILE_RES
0016 1 COMPLEX ZRESBUF
0017 1 DATA C/4900./
0018 DIMENSION F(3),V(3)
0019 DATA LN/21/,IRCHX/256/,EPSILON/1.E-5/
0020 CALL GAI(5,' NO OF CHANNELS = ',NCHAN)
0021 DO 10 N=1,NCHAN
0022 CALL TEXTI(5,'$CHANNEL',N)
0023 CALL GAA(5,'+ TIME SERIES INPUT FILE: ',FILE_RES(1,N))
0024 CALL TEXTI(5,'$CHANNEL',N)
0025 CALL GAR(5,'+ SNR(DB) = ',SNR_RES(N))
0026 CALL TEXTI(5,'$CHANNEL',N)
0027 CALL GAR(5,'+ SIGMA SNR(DB) = ',SIGSNR_RES(N))
0028
0029 10 CONTINUE
0030 CALL TEXT(5,' ')
0031 CALL TEXT(5,' CHANNEL FRQ. BINSZ SRATE STIME
NSAMP')
0032 DO 20 N=1,NCHAN
0033 CALL GTSHDR(LN,FILE_RES(1,N),IRCHX,FNS_RES(N),BSZ_RES(N),
SR_RES(N),TSSR,CFQ_RES(N),T_RES(N),FOS_RES(1,N),IERR)
0034 WRITE(5,1300)N,CFQ_RES(N),BSZ_RES(N),SR_RES(N),T_RES(N)
0035 ,FNS_RES(N)
0036 FOS_RES(1,N)=FOS_RES(1,N)*6076.
0037 FOS_RES(2,N)=FOS_RES(2,N)*6076.
0038 FOS_RES(3,N)=FOS_RES(3,N)*6076.
0039
0040 20 CONTINUE
0041 FORMAT(15,F12.3,F10.3,F10.3,F7.0,F10.0)
0042 CALL TEXT(5,' ')
0043 CALL GAR(5,' STARTING TARGET ESTIMATION TIME(SEC) = ',TSTRI_REF)
0044 CALL GAR(5,' ENDING TARGET ESTIMATION TIME(SEC) = ',TEND_REF)
0045 CALL GAI(5,' TIME SKIP BETWEEN ESTIMATIONS(SEC) = ',TSKIP)
0046 CALL GAI(5,' NO OF SAMPLES PER INTEGRATION = ',NSAMP)
0047 CALL GAI(5,' DEGREE OF POSITION VS TIME POLYNOMIALS = ',NDEG)
0048 CALL GAR(5,' SINC FUNCTION TRUNCATION FT = ',TRUNC)
0049 CALL GAR(5,' INITIAL X-COORDINATE(FT) = ',XINIT)
0050 CALL GAR(5,' INITIAL Y-COORDINATE(FT) = ',YINIT)
0051 CALL GAR(5,' INITIAL Z-COORDINATE(DEPTH-FT) = ',ZINIT)
0052 CALL GAR(5,' INITIAL SPEED(KTS) = ',VINIT)
0053 VINIT=VINIT*6076.029/3600.
0054 CALL GAR(5,' INITIAL HEADING(DEG) = ',BINIT)
0055 DO 30 N=0,NDEG
0056 SV(1,N)=0.
0057 SV(2,N)=0.

```

TRDIRV

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14-May-1985 12:14:28VAX-11 FORTRAN V3.2-37
DISKUSER1:CELDRIGEJTRDIRV1.FOR:21

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```
0058      SV(3,N)=0.
0059      CONTINUE
0060      SV(1,0)=XINIT
0061      SV(2,0)=YINIT
0062      SV(3,0)=ZINIT
0063      SV(1,1)=VINIT*SIND(BINIT)
0064      SV(2,1)=VINIT*COSD(BINIT)
0065      NSTATE=2*NDEG+2
0066      TEST=TSIRT_REF
0067      DO 1000 WHILE(TEST.LE.TEND_REF)
0068          CALL TRACK(TEST,ZZ,RR,IERR)
0069          IF(IERR.EQ.0)CALL TRERREF(IERR)
0070          CALL TRUPDAT(P,V,TEST)
0071          CALL TEXT(S,' ')
0072          CALL TEXTR(S,' TS(SEC) = ',TEST)
0073          CALL TEXTR(S,' X(TS) = ',F(1))
0074          CALL TEXTR(S,' Y(TS) = ',F(2))
0075          SPEED=SQRT(V(1)**2+V(2)**2)*3600./6076.
0076          COURSE=ATAN2D(V(1),V(2))
0077          CALL TEXTR(S,' SPEED(KTS) = ',SPEED)
0078          CALL TEXTR(S,' COURSE(DEG) = ',COURSE)
0079          CALL TRUPDAT(P,V,TEST+TSKIP)
0080          SV(1,0)=F(1)
0081          SV(2,0)=F(2)
0082          SV(3,0)=F(3)
0083          SV(1,1)=V(1)
0084          SV(2,1)=V(2)
0085          SV(3,1)=V(3)
0086          DO 40 K=2,NDEG
0087              SV(1,N)=0
0088              SV(2,N)=0
0089              SV(3,N)=0
0090          CONTINUE
0091          TEST=TEST+TSKIP
0092          CONTINUE
0093          STOP
0094          END
```

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```

0001 SURROUTINE TRUFAT(F,V,TIME)
0002 INCLUDE 'TRWRK.CMN'
0003 1 DOUBLE PRECISION Z,ZD,X,F,FX,BR,RHOSQ
0004 1 COMMON/TRWRK/SV(3,0:10),F,FX(31),BR(31),RHO,RHOSQ,Z(961),ITCOUNT,IS
0005 1 ,DX(31),X(31),COV(4,4),DETCOV
0006 1 INCLUDE 'TRCNTRL.CMN'
0007 1 COMMON/TRCNTRL/NDEG,NSAMP,TSTRT_REF,TEND_REF,NSKIP,MAXIT,EPSILON
0008 1 ,NCHAN,TRUNC,NSTATE,C
0009 DIMENSION F(3),V(3)
0010 TI=TIME-TS
0011 DO 10 N=1,3
0012 F(N)=SV(N,NDEG)
0013 V(N)=0.
0014 10 CONTINUE
0015 DO 30 M=NDEG-1,0,-1
0016 MF1=MF1
0017 DO 20 N=1,3
0018 F(N)=F(N)*TI+SV(N,M)
0019 V(N)=V(N)*TI+(MF1)*SV(N,MF1)
0020 20 CONTINUE
0021 30 CONTINUE
0022 RETURN
0023 END

```

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```

0001 SUBROUTINE TRACK(IT,Z,B,IERR)
0002 INCLUDE 'TRCNTRL.CMN'
0003 COMMON/TRCNTRL/NDEG,NSAMP,ISTRT_REF,TEND_REF,NSKIP,MAXIT,EPSILON
0004 ,NCHAN,TRUNC,NSTATE,C
0005 INCLUDE 'TRWOK.CMN'
0006 DOUBLE PRECISION Z,ZD,X,F,FX,BB,RHOSQ
0007 COMMON/TRWOK/SV(3,0:10),F,FX(31),BB(31),RHQ,RHOSQ,ZZ(961),ITCOUNT,TS
0008 ,DX(31),X(31),COV(4,4),DETCOV
0009 INCLUDE 'TRTMS.CMN'
0010 COMMON/TRTMS/FILE_REF(32),CFQ_REF,BSZ_REF,SR_REF,T_REF,POS_REF(3)
0011 ,FNS_REF,SNR_REF,SIGSNR_REF
0012 ,FILE_RES(32,5),CFQ_RES(5),BSZ_RES(5),SR_RES(5),T_RES(5)
0013 ,FOS_RES(3,5),FNS_RES(5),SNR_RES(5),SIGSNR_RES(5),COH(5),BIAS(5)
0014 ,SIGRQCOH(5),ZRESRUF(0:4200),T_BUF
0015 BYTE FILE_REF,FILE_RES
0016 COMPLEX ZRESRUF
0017 DOUBLE PRECISION FT,Z,DET,AWK,BWK,B,RFACT,ZV,ZW
0018 DIMENSION Z(NSTATE,NSTATE),B(NSTATE)
0019 DIMENSION F(3),V(3),U(3),UN(3),DP(3),DI(3),AWK(31),BWK(31)
0020 COMPLEX ZREF(4000),ZRES(4000),DZRESDT,TWOPIJ
0021 COMPLEX FTC,ZEXP,DZRESRDT
0022 COMPLEX*16 ZU,ZDW(33),ZDU(33),DFACT,CDOT
0023 DATA NBUFRF/4000/,NBUFRFES/4500/,TWOPIJ/(0.,6.283185307)/,LN/21/
0024 DATA IRCMX/256/,MXLOOP/20/,ALN10/.230258509/
0025 TS=TT
0026 SFACF=(NSAMP-1.)/(NSAMP**2+NSAMP)
0027 DO 5 I=0,NDEG
0028   X(I+1)=SV(1,I)
0029   X(I+NDEG+2)=SV(2,I)
0030 CONTINUE
0031 C
0032   X(NSTATE)=SV(3,0)
0033   IL0OF=0
0034   GWGOLD=1.E30
0035   DO 14 I=1,NSTATE
0036     B(I)=0.
0037     DO 14 J=1,I
0038       Z(I,J)=0.
0039       Z(J,I)=0.
0040 CONTINUE
0041 GWG=0.
0042 DO 700 NREF=1,NCHAN-1
0043   CALL GTSDIR(LN,FILE_RES(1,NREF),IRCMX,FNS_REF,BSZ_REF,SR_REF,TSSR,
0044   ,CFQ_REF,T_REF,POS_REF,IERR)
0045   FOS_REF(1)=FOS_REF(1)*6076.
0046   FOS_REF(2)=FOS_REF(2)*6076.
0047   FOS_REF(3)=FOS_REF(3)*6076.
0048   SNR_REF=SNR_RES(NREF)
0049   SIGSNR_REF=SIGSNR_RES(NREF)
0050   FCCALC=1./(1.00001+10.**((-SNR_REF/10.))
0051   DELTA_T=1./SR_REF
0052   TINT=NSAMP*DELTA_T
0053   NMIN=MAX(TRUNC,(TT-TINT/2.-T_REF)*SR_REF)
0054   CALL GTSDAT(LN,FILE_RES(1,NREF),IRCMX,NMIN,NSAMP,ZREF,IERR)
0055   ZU=CDOT(ZREF,ZREF,NSAMP)
0056   DO 700 NC=NREF+1,NCHAN
0057     COH(NC)=FCCALC/(1.00001+10.**((-SNR_RES(NC)/10.))

```

```
0058      BIAS(NC)=(1.-COH(NC))*2*(1.+2.*COH(NC)/NSAMP)/NSAMP
0059      SIGSREF=(ALN10*COH(NC))/(1.+10.*(SNR_REF/10.))*SIGSNR_REF)**2
0060      SIGSREF=(ALN10*COH(NC))/(1.+10.*(SNR_RES(NC)/10.))
0061      *SIGSNR_RES(NC)**2
0062      SIGSCOH(NC)=SIGSREF+SIGSRES
0063      DO 40 N=1,NSATE
0064      ZUW(N)=(0.,0.)
0065      ZIU(N)=(0.,0.)
0066      CONTINUE
0067      F=0.
0068      ZU=(0.,0.)
0069      ZW=(0.,0.)
0070      ZNUM=(0.,0.)
0071      TF=T_REF+MIN/SR_REF
0072      CALL TRUFDAI(F,V,TF)
0073      TG=TF-ABSRVSH(1.,F,-1.,FOS_REF,3)/C
0074      ANDRES=0.
0075      ZEXP=(1.,0.)
0076      CALL TOT1(TAUBAR,TW,TG,TF,FOS_REF,FOS_RES(1,NC),EFSILON,ERR)
0077      DO 600 NRF=1,NSAMP
0078      CALL TRUFDAI(F,V,TW)
0079      CALL RSMADD(UO,1.,F,-1.,FOS_REF,3)
0080      CALL RSMADD(UN,1.,F,-1.,FOS_RES(1,NC),3)
0081      DO=ABSRV(UO,3)
0082      IN=ABSRV(UN,3)
0083      CALL RVSCM(UO,1./DO,UO,3)
0084      CALL RVSCM(UN,1./IN,UN,3)
0085      CALL XN-DXN(TAUBAR,NC,ZRES(NRF),DZRESDT)
0086      ZRES(NRF)=ZRES(NRF)*ZEXP
0087      DFACT=TWOFI*CFQ_RES(NC)*ZRES(NRF)+ZEXP*DZRESDT
0088      DSNDTK=RDOT(V,UN,3)/C
0089      DSODTK=RDOT(V,UO,3)/C
0090      RFACT=(1.DO+DSNDTK)/(1.DO+DSODTK)
0091      DO 60 NS=1,NSATE
0092      CALL DFDT(DP,TW,NS)
0093      DSNDTH=RDOT(UN,DP,3)/C
0094      DSODTH=RDOT(UO,DP,3)/C
0095      DTAUDTH=-DSODTH*RFACT+DSNDTH
0096      DZRESDTH=DFACT*DTAUDTH
0097      ZUW(NS)=ZUW(NS)+DZRESDTH*CONJG(ZRES(NRF))
0098      ZIU(NS)=ZIU(NS)+DZRESDTH*CONJG(ZREF(NRF))
0099      CONTINUE
0100      ZW=ZW+ZRES(NRF)*CONJG(ZRES(NRF))
0101      ZU=ZU+ZREF(NRF)*CONJG(ZRES(NRF))
0102      TAU_LAST=TAUBAR
0103      TG=TW+DELTA_T
0104      TF=TF+DELTA_T
0105      CALL TOT1(TAUBAR,TW,TG,TF,FOS_REF,FOS_RES(1,NC),EFSILON,ERR)
0106      FT=CFQ_RES(NC)*(TAUBAR-TAU_LAST)
0107      FTC=MOD(FT,1.DO)
0108      ZEXP=ZEXP*CEXP(TWOFI*FTC)
0109      CONTINUE
0110      RHOSQ=ZU*CONJG(ZU)
0111      RHOSQ=RHOSQ/(ZV*ZW)
0112      F=LOG((COH(NC)+BIAS(NC))/RHOSQ)
0113      WF=(COH(NC)+BIAS(NC))*2/SIGSCOH(NC)
0114      GWG=GWG+WF*F**2
```

```

0115      DO 650 NS=1,NSTATE
0116          FX(NS)=2.*DREAL(ZIW(NS))/ZW-2.*DREAL(ZIU(NS)/CONJG(ZU))
0117      CONTINUE
0118      DO 660 I=1,NSTATE
0119          R(I)=R(I)+WF*FX(I)
0120      DO 670 J=1,I
0121          Z(I,J)=Z(I,J)+FX(I)*FX(J)*WF
0122          Z(J,I)=Z(I,J)
0123      CONTINUE
0124      CONTINUE
0125      CALL DMINV(Z,NSTATE,DEL,AWK,BWK)
0126      CHK=0.
0127      DO 720 I=1,NSTATE
0128          DX(I)=0.
0129          DO 710 K=1,NSTATE
0130              DX(I)=DX(I)-Z(I,K)*B(K)
0131          CONTINUE
0132          CHK=CHK+B(I)*R(I)
0133      CONTINUE
0134      ARSB=SQRT(CHK)
0135      DO 725 I=1,NSTATE
0136          DX(I)=-.5*GWG*H(I)/ARSB
0137      CONTINUE
0138      IF(GWG.GF.GWGOLD)GO TO 810
0139      DO 730 I=0,NDEG
0140          X(I+)=X(I+)+DX(I+1)
0141          X(I+NDEG+2)=X(I+NDEG+2)+DX(I+NDEG+2)
0142          SV(1,I)=X(I+1)
0143          SV(2,I)=X(I+NDEG+2)
0144      CONTINUE
0145      SV(3,0)=X(NSTATE)
0146      GWGOLD=GWG
0147      ILOOP=ILOOP+1
0148      IF(ILOOP.LE.MXLOOP)GO TO 10
0149      IERR=1
0150      RETURN
0151      IERR=0
0152      RETURN
0153      END
0154

```

40 1103 1700 17:33:20
15-Oct-1984 12:35:21

VMAT11.FUNTRM V3.2.23/
DISK\$USER1:ELDRINGE\TRUTILS1.FOR:2

```

0001 SUBROUTINE IOI1(T1,TW,TG,TO,FO,F1,EPSLN,ERR)
0002
0003 C THIS SUBROUTINE SOLVES FOR THE RECEIVER TIME, T1, ON CHANNEL 1,
0004 C WHICH CORRESPONDS TO THE RECEIVER TIME, TO, ON CHANNEL 0. THE
0005 C CURRENT VALUE OF THE STATE VECTOR IS USED TO MAKE THESE CALCU-
0006 C LATIONS. FO AND F1 ARE THE POSITION VECTORS OF STATIONS 0 AND 1,
0007 C RESPECTIVELY. TG IS THE INITIAL GUESS AT THE INTERMEDIATE WATER
0008 C TIME, TW. THE SUBROUTINE RETURNS T1, TW, AND ERR. THE VARIABLE
0009 C EPSLN IS THE TOLERANCE IN SECONDS FOR THE CONVERGENCE CRITERION.
0010 C THE SUBROUTINE WILL ITERATE AT MOST 20 TIMES IF CONVERGENCE IS
0011 C NOT MET. IT WILL THEN SIGNAL WITH AN ERROR MESSAGE TO THE OPERATOR
0012 C TERMINAL AND SET ERR = 1. OTHERWISE THE SUBROUTINE WILL RETURN WITH
0013 C ERR = 0.
0014 C
0015 INCLUDE 'ELDRINGE\TRCNTRL.CMN'
0016 COMMON/TRCNTRL/NDG,NSAMP,TSTRT_REF,TEND_REF,NSKIP,MAXIT,EPSILON
0017 C ,NCHAN,TRUNC,NSTATE,C
0018 C DIMENSION F(3),FO(3),F1(3),V(3),U(3)
0019 C ERR=0.
0020 C TL=TG
0021 C DO 10 N=1,20
0022 C   CALL TRUFDT(F,V,TL)
0023 C   G=ABSRVSM(1.,F,-1.,FO,3)
0024 C   CALL RSHADD(U,1.,F,-1.,FO,3)
0025 C   CALL RVSCM(U,1./G,U,3)
0026 C   F=G+C*(TL-TO)
0027 C   FF=C+RDOT(U,V,3)
0028 C   DELT=-F/FF
0029 C   TL=TL+DELT
0030 C   IF (ABS(DELT).LE.EPSLN)GO TO 20
0031 C
0032 C CONTINUE
0033 C CALL TEXT(5,' NO CONVERGENCE AFTER 20 ITERATIONS' )
0034 C ERR=1.
0035 C RETURN
0036 C
0037 C NOW SET TW = TL AND PLUG THIS VALUE INTO THE RECEIVER TIME
0038 C EQUATION FOR CHANNEL 1 TO OBTAIN T1
0039 C
0040 C TW=TL
0041 C CALL TRUFDT(F,V,TW)
0042 C G1=ABSRVSM(1.,F,-1.,F1,3)
0043 C T1=TW+G1/C
0044 C RETURN
0045 C END

```

FUNCTION ABSRV(F,N)

0001
0002 C
0003 C
0004 C
0005 CTHIS FUNCTION COMPUTES THE EUCLIDEAN NORM OF AN N-DIMENSIONAL
REAL VECTOR F

0006 DIMENSION F(1)

0007 ABSRV=0.

0008 DO 10 I=1,N

0009 ABSRV=ABSRV+F(I)*F(I)

0010 CONTINUE

0011 ABSRV=SQRT(ABSRV)

0012 RETURN

0013 END

10


```

0001 FUNCTION ABSRVSM(S1,F1,S2,F2,N)
0002 C
0003 THIS FUNCTION COMPUTES THE ABSOLUTE VALUE OF THE REAL
0004 VECTOR
0005 C      S1*F1+S2*F2
0006 C WHERE S1 AND S2 ARE REAL SCALORS AND F1 AND F2 ARE REAL
0007 C N-DIMENSIONAL VECTORS.
0008 C
0009 DIMENSION F1(1),F2(1)
0010 ABSRVSM=0.
0011 DO 10 I=1,N
0012     T=S1*F1(I)+S2*F2(I)
0013     ABSRVSM=ABSRVSM+T*T
0014 C CONTINUE
0015 ABSRVSM=SQRT(ABSRVSM)
0016 RETURN
0017 END

```

10

```

0001 SUBROUTINE RVSCM(F,S1,F1,N)
0002 C
0003 C THIS SUBROUTINE PERFORMS A REAL SCALAR MULTIPLY
0004 C F=S1*F1
0005 C WHERE F,F1 ARE N-DIMENSIONAL REAL VECTORS AND S1 IS A
0006 C REAL SCALAR.
0007 C
0008 C DIMENSION F(1),F1(1)
0009 C DO 10 I=1,N
0010 C F(I)=S1*F1(I)
0011 C CONTINUE
0012 C RETURN
0013 C END

```

```

0002 SURROUTINE FCTSDAT(LN,IR,IFIL,IRCMX,NMIN,NSAMP,Z,IERR)
0003 ENTRY FTSDAT(LN,DFIL,IRCMX,NMIN,NSAMP,Z,IERR)
0004
0005 THIS SUBROUTINE WRITES COMPLEX DATA TO FILES COMPATIBLE WITH
0006 BRGENT. THE PROGRAM WRITES THE FIRST NSAMP VALUES
0007 OF THE COMPLEX ARRAY Z INTO NSAMP COMPLEX
0008 SAMPLES STARTING WITH THE ANMIN'TH SAMPLE IN THE FILE.
0009
0010 THIS SUBROUTINE ASSUMES IRCMX COMPLEX DFT'S PER RECORD.
0011
0012 IMPLICIT COMPLEX (Z)
0013 DIMENSION Z(1),ZBUF(256)
0014 BYTE DFIL(1)
0015 DATA IHEADER/1/,MAXZBUF/256/
0016
0017 IF(IRCMX.GT.MAXZBUF)GO TO 30
0018 CALL CLOSE(LN)
0019 OPEN(UNIT=LN,FILE=DFIL,STATUS='UNKNOWN',RECL=IRCMX*2,ACCESS='DIRECT',
0020 IERR=0
0021 BLOCKSIZE=IRCMX*8,ASSOCIATEVARIABLE=IR)
0022 IERR=0
0023 IRCMX=IRCMX
0024 ANMIN=NMIN+1
0025 ISREC=NMIN/RCMX+IHEADER+1
0026 ANMAX=NMIN+NSAMP
0027 IEREC=(ANMAX-1)/RCMX+IHEADER+1
0028 ADATA1=ANMIN
0029 N1=0
0030 IR=ISREC
0031 DO 10 IREC=ISREC,IEREC
0032 IRUF1=AMOD(ADATA1-1,RCMX)+1
0033 ADATA2=AMIN1((IEREC-IHEADER)*RCMX,ANMAX)
0034 IRUF2=AMOD(ADATA2-1,RCMX)+1
0035 IF(IRUF2-IRUF1+1.EQ.IRCMX)GO TO 4
0036 READ(LN,IR,ERR=3)(ZBUF(I),I=1,IRCMX)
0037 IR=IEREC
0038 DO 5 I=IRUF1,IRUF2
0039 N1=N1+1
0040 ZBUF(I)=Z(N1)
0041 CONTINUE
0042 WRITE(LN,IR,ERR=20)(ZBUF(I),I=1,IRCMX)
0043 ADATA1=ADATA1+IRUF2-IRUF1+1
0044 CONTINUE
0045 CALL CLOSE(LN)
0046 RETURN
0047 CALL TEXT(5,' FTSDAT/FCTSDAT: ERROR WRITING FILE: ')
0048 IERR=1
0049 CALL CLOSE(LN)
0050 RETURN
0051 CALL TEXT(5,' FTSDAT/FCTSDAT:IRCMX EXCEEDS MAXZBUF, IRCMX = ',IRCMX)
0052 CALL TEXT(5,' CURRENT VALUE OF MAXZBUF = ',MAXZBUF)
0053 CALL TEXT(5,' IF YOU INCREASE MAXZBUF,MAKE SURE YOU INCREASE THE')
0054 CALL TEXT(5,' DIMENSION OF ZBUF ACCORDINGLY.')
0055 STOP
0056 END

```

```

0002 SUBROUTINE GCTSDAT(LN,DFTFIL,IRCMX,NMIN,NSAMP,Z,IERR)
0003 ENTRY GTSB DAT(LN,DFTFIL,IRCMX,NMIN,NSAMP,Z,IERR)
0004
0005 THIS SUBROUTINE OBTAINS COMPLEX DATA FROM FILES GENERATED
0006 RECENT. THE PROGRAM RETURNS WITH THE FIRST NSAMP VALUES
0007 OF THE COMPLEX ARRAY Z LOADED FROM THE NSAMP COMPLEX
0008 SAMPLES STARTING WITH THE ANMIN'TH SAMPLE IN THE FILE. THE FIRST
0009 SAMPLE IN THE FILE IS NUMBERED 0 TO THE USER
0010
0011
0012 THIS SUBROUTINE ASSUMES IRCMX COMPLEX DFT'S PER RECORD.
0013
0014 IMPLICIT COMPLEX (Z)
0015 DIMENSION Z(1),ZBUF(256)
0016 BYTE DFTFIL(1)
0017 DATA IHEADER/1/,MAXZBUF/256/
0018
0019 IF (IRCMX.GT.MAXZBUF)GO TO 30
0020 CALL CLOSE(LN)
0021 OPEN(UNIT=LN,FILE=DFTFIL,STATUS='OLD',RECL=IRCMX*2,ACCESS='DIRECT',
0022 BLOCKSIZE=IRCMX*8,ASSOCIATE VARIABLE=IR)
0023 IERR=0
0024 IRCMX=IRCMX
0025 ANMIN=NMIN+1
0026 ISREC=NMIN/RCMX+IHEADER+1
0027 ANMAX=NMIN+NSAMP
0028 IEREC=(ANMAX-1)/RCMX+IHEADER+1
0029 IR=ISREC
0030 ADATA1=ANMIN
0031 N1=0
0032 DO 10 IREC=ISREC,IEREC
0033 READ(LN,IR,ERR=20)(ZBUF(1),I=1,IRCMX)
0034 IRUF1=AMOD(ADATA1-1,RCMX)+1
0035 ADATA2=AMIN1((IEREC-IHEADER)*RCMX,ANMAX)
0036 IRUF2=AMOD(ADATA2-1,RCMX)+1
0037 DO 5 I=IRUF1,IRUF2
0038 N1=N1+1
0039 Z(N1)=ZBUF(1)
0040
0041 CONTINUE
0042 ADATA1=ADATA1+IRUF2-IRUF1+1
0043
0044 CONTINUE
0045 CALL CLOSE(LN)
0046 RETURN
0047
0048 CALL TEXT(5,' GTSB DAT/GCTSDAT: ERROR READING FILE: ')
0049 IERR=1
0050 CALL CLOSE(LN)
0051 RETURN
0052
0053 CALL TEXT(5,' GTSB DAT/GCTSDAT: IRCMX EXCEEDS MAXZBUF. IRCMX = ',IRCMX)
0054 CALL TEXT(5,' CURRENT VALUE OF MAXZBUF = ',MAXZBUF)
0055 CALL TEXT(5,' IF YOU INCREASE MAXZBUF, MAKE SURE YOU INCREASE THE ')
0056 CALL TEXT(5,' DIMENSION OF ZBUF ACCORDINGLY. ')
0057 STOP
0058 END

```

```

=====
C-----
0001 SURROUTINE FTSHDR(LN,DFTFIL,IRCMX,TDFTS,BINSZ,SSR,TSSR,
0002 BINFRO,TDEL,FVEC,IERR)
0003 BYTE DFTFIL(1)
0004 DIMENSION FVEC(3)
0005
0006
0007 BINSZ = BIN SIZE(HZ)
0008 BINFRO = CENTER FREQUENCY(HZ) OF 1' TH BIN
0009 SSR = SPECTRAL SAMPLE RATE
0010 TSSR = TIME SERIES SAMPLE RATE(HZ)
0011 IRCMX = COMPLEX RECORD SIZE (256)
0012 TDFTS= TOTAL # OF DFT COEFF.S (samples)
0013 TDEL = channel delay
0014 FVEC(3) = POSITION VECTOR OF SENSOR
0015
0016
0017
0018
0019
0020 CALL CLOSE(LN)
0021 OPEN(UNIT=LN,FILE=DFTFIL,STATUS='UNKNOWN',RECL=IRCMX*2,ACCESS='DIRECT',
0022 BLOCKSIZE=IRCMX*8,ASSOCIATEVARIABLE=IR)
0023
0024 IR=1
0025 IERR=0
0026 WRITE(LN,IR,IERR=900)IRCMX,TDFTS,BINSZ,SSR,TSSR,BINFRO,TDEL,FVEC
0027 GO TO 999
0028
0029 CALL TEXT(5,' FTSHDR: WRITE ERROR')
0030 IERR=1
0031 ! write error
0032 CONTINUE
0033 CALL CLOSE(LN)
0034 RETURN
0035 END
=====

```

```

0001 SUBROUTINE GTSHDR(LN,DFTFIL,IRCMX,TDFTS,BINSIZE,SSR,TSSR,
0002 BINFRO,TDEL,FVEC,IERR)
0003 BYTE DFTFIL(1)
0004 DIMENSION FVEC(3)
0005
0006 C BINSIZE = BIN SIZE(HZ)
0007 C BINFRO = CENTER FREQUENCY(HZ) OF 1' TH BIN
0008 C SSR = SPECTRAL SAMPLE RATE
0009 C TSSR = TIME SERIES SAMPLE RATE(HZ)
0010 C IRCMX = COMPLEX RECORD SIZE (256)
0011 C TDFTS= TOTAL # OF DFT COEFF.S (samples)
0012 C TDEL = channel delay
0013 C FVEC(3) = POSITION VECTOR OF SENSOR
0014 C
0015 C =====
0016 C
0017 C
0018 CALL CLOSE(LN)
0019 OPEN(UNIT=LN,FILE=DFTFIL,STATUS='OLD',RECL=IRCMX*2,ACCESS='DIRECT',
0020 BLOCKSIZE=IRCMX*8,ASSOCIATEVARIABLE=IR)
0021 C
0022 IR=1
0023 IERR=0
0024 READ(LN,IR,ERR=900)IRCMX,TDFTS,BINSIZE,SSR,TSSR,BINFRO,TDEL,FVEC
0025 GO TO 999
0026 C
0027 900 CALL TEXT(5,' GTSHDR: READ ERROR')
0028 IERR=1
0029 ! read error
0029 CONTINUE
0030 CALL CLOSE(LN)
0031 RETURN
0032 END

```

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VAX-11 PUKIKAN V3.2-37
DISK\$USER1:TELNRIDGEJTRUTILS1.FOR;2 Page 14

SUBROUTINE TRERKEP(I)

THIS SUBROUTINE IS A DUMMY(AS IS ITS AUTHOR!).

RETURN
END

0001 C
0002 C
0003 C
0004 C
0005
0006

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 15-Oct-1984 12:35:21

```

0001 SUBROUTINE DF0TH(D,I1,K)
0002 DIMENSION D(3)
0003 INCLUDE 'TRCNTRL.CMN'
0004 1 COMMON/TRCNTRL/NDEG,NSAMP,TSTRT_REF,TEND_REF,NSKIP,MAXIT,EPSILON
0005 1 ,NCHAN,TRUNC,NSTATE,C
0006 1 INCLUDE 'TRWORK.CMN'
0007 1 DOUBLE PRECISION ZZ,DX,FX,BB,RHOSQ
0008 1 COMMON/TRWORK/SV(3,0:10),FX(31),RH0,RHOSQ,ZZ(961),ITCOUNT,TS
0009 1 ,IX(31),X(31),COV(4,4),DETCOV
0010 K1=K-1
0011 T=T1-TS
0012 IF(K1.LE.NDEG)THEN
0013   D(1)=T**K1
0014   D(2)=0.
0015   D(3)=0.
0016 ELSE
0017   IF(K1.LE.2*NDEG+1)THEN
0018     D(1)=0.
0019     D(2)=T**(K1-NDEG-1)
0020     D(3)=0.
0021   ELSE
0022     D(1)=0.
0023     D(2)=0.
0024     D(3)=1.
0025   ENDIF
0026   ENDIF
0027   RETURN
0028   END

```



```
0001 FUNCTION DSINCDT(X)
0002 DOUBLE PRECISION DSINCDT,X
0003 IF (X.EQ.0.D0) THEN
0004   DSINCDT=0.D0
0005 ELSE
0006   DSINCDT=COS(X)/X-SIN(X)/X**2
0007 ENDIF
0008 RETURN
0009 END
```

```
0001 FUNCTION SINC(X)
0002 DOUBLE PRECISION SINC,X
0003 IF(X.EQ.0.D0)THEN
0004   SINC=1.D0
0005 ELSE
0006   SINC=SIN(X)/X
0007 ENDIF
0008 RETURN
0009 END
```

```

0001 SUBROUTINE XN_DYN(IBAR,N,ZXN,DZXNDT)
0002 INCLUDE 'ELDRIDGEJTRTMS.CMN'
0003 COMMON/TRTMS/FILE_REF(32),CFQ_REF,BSZ_REF,SR_REF,T_REF,POS_REF(3)
0004 .FNS_REF,SNR_REF,SIGSNR_REF
0005 .FILE_RES(32,5),CFQ_RES(5),BSZ_RES(5),SR_RES(5),T_RES(5)
0006 .POS_RES(3,5),FNS_RES(5),SNR_RES(5),SIGSNR_RES(5),COH(5),BIAS(5)
0007 .SIGSCOH(5),ZRESBUF(0:4200),T_BUF
0008 .BYTE FILE_REF,FILE_RES
0009 COMPLEX ZRESBUF
0010 INCLUDE 'ELDRIDGEJTRCNTRL.CMN'
0011 .COMMON/TRCNTRL/NIEG,NSAMP,TSTRT_REF,TEND_REF,NSKIP,MAXIT,EPSILON
0012 .NCHAN,TRUNC,NSTATE,C
0013 COMPLEX ZXN,DZXNDT,ZBUF(100)
0014 COMPLEX*16 DZXN,DZXNDT
0015 DOUBLE PRECISION PI,SAMPN,ARG,SIGMA,SINC,DSINCDT
0016 DATA PI/3.141592653589793D0/,LN/21/,IRCMX/256/
0017 TRUNC2=2*TRUNC
0018 SIGMA=PI*SR_RES(N)
0019 ID=IBAR-T_RES(N)
0020 SAMPN=ID*SR_RES(N)
0021 NMIN=SAMPN-TRUNC+1
0022 NMAX=NMIN+TRUNC2-1
0023 NPTS=NMAX-NMIN+1
0024 CALL GTSDAT(LN,FILE_RES(1,N),IRCMX,NMIN,NPTS,ZBUF,IERR)
0025 NN=1
0026 DZXN=(0.,0.)
0027 DDZXNDT=(0.,0.)
0028 DO 10 NS=NMIN,NMAX
0029 ARG=(SAMPN-NS)*PI
0030 DZXN=DZXN+SINC(ARG)*ZBUF(NN)
0031 DDZXNDT=DDZXNDT+DSINCDT(ARG)*ZBUF(NN)
0032 NN=NN+1
0033 CONTINUE
0034 ZXN=DZXN
0035 DDZXNDT=SIGMA*DDZXNDT
0036 RETURN
0037 END

```

```

0001 SUBROUTINE CSMAAD(Z,S1,Z1,S2,Z2,N)
0002 C
0003 C THIS SUBROUTINE COMPUTES THE VECTOR SUM
0004 C Z=S1*Z1+S2*Z2
0005 C WHERE Z,Z1,AND Z2 ARE COMPLEX N-DIMENTIONAL VECTORS AND
0006 C S1 AND S2 ARE COMPLEX SCALARS.
0007 C
0008 C COMPLEX Z,Z1,Z2,S1,S2
0009 C DIMENSION Z(1),Z1(1),Z2(1)
0010 C DO 10 I=1,N
0011 C   Z(I)=S1*Z1(I)+S2*Z2(I)
0012 C CONTINUE
0013 C RETURN
0014 C
0015 10

```

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```
0001 SUBROUTINE RSHADD(P,S1,P1,S2,P2,N)
0002 C
0003 C THIS SUBROUTINE COMPUTES THE VECTOR SUM
0004 C P=S1*P1+S2*P2
0005 C WHERE P,P1,AND P2 ARE REAL N-DIMENSIONAL VECTORS AND S1 AND S2
0006 C ARE REAL SCALARS
0007 C
0008 C DIMENSION P(1),P1(1),P2(1)
0009 DO 10 I=1,N
0010 P(I)=S1*P1(I)+S2*P2(I)
0011 CONTINUE
0012 RETURN
0013 END
```

COMPLEX FUNCTION CDOT*16(Z1,Z2,N)

THIS FUNCTION COMPUTES THE COMPLEX DOT PRODUCT BETWEEN THE
 N-DIMENSIONAL COMPLEX VECTORS Z1 AND Z2

COMPLEX Z1,Z2

DIMENSION Z1(1),Z2(1)

CDOT=(0.,0.)

DO 10 I=1,N

CDOT=CDOT+Z1(I)*CONJG(Z2(I))

CONTINUE

RETURN

END

0001 C
 0002 C
 0003 C
 0004 C
 0005 C
 0006
 0007
 0008
 0009
 0010 10
 0011
 0012
 0013

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FUNCTION RDOT(V1,V2,N)

THIS FUNCTION COMPUTES THE DOT PRODUCT BETWEEN THE REAL
 N-DIMENSIONAL VECTORS V1 AND V2.

DIMENSION V1(1),V2(1)

RDOT=0.

DO 10 I=1,N

RDOT=RDOT+V1(I)*V2(I)

CONTINUE

RETURN

END

0001
 0002 C
 0003 C
 0004 C
 0005 C
 0006
 0007
 0008
 0009
 0010 10
 0011
 0012

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15-Oct-1984 12:35:21

```
0001 SUBROUTINE CVSCH(Z,S1,Z1,N)
0002 C
0003 C THIS SUBROUTINE PERFORMS A COMPLEX SCALAR MULTIPLY
0004 C Z=S1*Z1
0005 C WHERE Z,Z1 ARE N-DIMENSIONAL COMPLEX VECTORS AND S1 IS A
0006 C COMPLEX SCALAR.
0007 C
0008 COMPLEX Z,Z1,S1
0009 DIMENSION Z(1),Z1(1)
0010 DO 10 I=1,N
0011 Z(I)=S1*Z1(I)
0012 CONTINUE
0013 RETURN
0014 END
```


28-May-1985 17:33:20
15-Oct-1984 12:35:21

```

=====
0001 SUBROUTINE GRTSDAT(LN,DFTFIL,IRCMX,NMIN,NSAMP,B,IERR)
0002
0003
0004 C THIS SUBROUTINE OBTAINS REAL DATA FROM FILES GENERATED BY SUBROUTINE
0005 C PRTSDAT. THE PROGRAM RETURNS WITH THE FIRST NSAMP VALUES
0006 C OF THE REAL ARRAY B LOADED FROM THE NSAMP REAL
0007 C SAMPLES STARTING WITH THE ANMIN'TH SAMPLE IN THE FILE. THE FIRST
0008 C SAMPLE IN THE FILE IS NUMBERED 0 TO THE USER
0009 C
0010 C
0011 C THIS SUBROUTINE ASSUMES IRCMX REAL DATA SAMPLES PER RECORD.
0012 C
0013 DIMENSION B(1),BUF(256)
0014 BYTE DFTFIL(1)
0015 DATA IHEADER/1,MAXBUF/256/
0016
0017 IF (IRCMX.GT.MAXBUF)GO TO 30
0018 CALL CLOSE(LN)
0019 OPEN(UNIT=LN,FILE=DFTFIL,STATUS='OLD',RECL=IRCMX,ACCESS='DIRECT',
0020 BLOCKSIZE=IRCMX*4,ASSOCIATEVARIABLE=IR)
0021 IERR=0
0022 IRCMX=IRCMX
0023 ANMIN=NMIN+1
0024 ISREC=NMIN/RCMX+IHEADER+1
0025 ANMAX=NMIN+NSAMP
0026 IEREC=(ANMAX-1.)/RCMX+IHEADER+1
0027 IR=ISREC
0028 ADATA1=ANMIN
0029 N1=0
0030 DO 10 IREC=ISREC,IEREC
0031 READ(LN,IR,ERR=20)(BUF(I),I=1,IRCMX)
0032 IBUF1=AMOD(ADATA1-1.,RCMX)+1
0033 ADATA2=AMIN1((IEREC-IHEADER)*RCMX,ANMAX)
0034 IBUF2=AMOD(ADATA2-1.,RCMX)+1
0035 DO 5 I=IBUF1,IBUF2
0036 N1=N1+1
0037 B(N1)=BUF(I)
0038
0039 CONTINUE
0040 ADATA1=ADATA1+IBUF2-IBUF1+1
0041
0042 CONTINUE
0043 CALL CLOSE(LN)
0044 RETURN
0045 IERR=1
0046 CALL CLOSE(LN)
0047 RETURN
0048 CALL TEXT(5,' GRTSDAT:IRCMX EXCEEDS MAXBUF. IRCMX = ',IRCMX)
0049 CALL TEXT(5,' CURRENT VALUE OF MAXBUF = ',MAXBUF)
0050 CALL TEXT(5,' IF YOU INCREASE MAXBUF,MAKE SURE YOU INCREASE THE')
0051 CALL TEXT(5,' DIMENSION OF BUF ACCORDINGLY.')
0052 STOP
END

```

```

=====
SUBROUTINE PRSTDAT(LN,DFTFIL,IRCMX,NMIN,NSAMP,R,IERR)
C
C THIS SUBROUTINE WRITES REAL DATA TO FILES COMPATIBLE WITH
C GRSTDAT. THE PROGRAM WRITES THE FIRST NSAMP VALUES
C OF THE REAL ARRAY B INTO NSAMP REAL
C SAMPLES STARTING WITH THE ANMIN'TH SAMPLE IN THE FILE.
C
C THIS SUBROUTINE ASSUMES IRCMX REAL SAMPLES PER RECORD.
C
C DIMENSION B(1),BUF(256)
C BYTE DFTFIL(1)
C DATA IHEADER/1/,MAXBUF/256/
C
C IF (IRCMX.GT.MAXBUF)GO TO 30
C CALL CLOSE(LN)
C OPEN(UNIT=LN,FILE=DFTFIL,STATUS='UNKNOWN',RECL=IRCMX,ACCESS='DIRECT',
C BLOCKSIZE=IRCMX*4,ASSOCIATEVARIABLE=IR)
C IERR=0
C RCMX=IRCMX
C ANMIN=NMIN+1
C ISREC=NMIN/RCMX+IHEADER+1
C ANMAX=NMIN+NSAMP
C IREC=(ANMAX-1.)/RCMX+IHEADER+1
C ADATA1=ANMIN
C N1=0
C IR=ISREC
C DO 10 IREC=ISREC,IREC
C   IRUF1=AMOD(ADATA1-1.,RCMX)+1
C   ADATA2=AMIN1((IREC-IHEADER)*RCMX,ANMAX)
C   IRUF2=AMOD(ADATA2-1.,RCMX)+1
C   IF (IRUF2-IRUF1+1.EQ.IRCMX)GO TO 4
C   READ(LN,IR,ERR=3)(BUF(I),I=1,IRCMX)
C   IR=IREC
C   DO 5 I=IRUF1,IRUF2
C     N1=N1+1
C     BUF(I)=B(N1)
C   CONTINUE
C   WRITE(LN,IR,ERR=20)(BUF(I),I=1,IRCMX)
C   ADATA1=ADATA1+IRUF2-IRUF1+1
C CONTINUE
C CALL CLOSE(LN)
C RETURN
C CALL TEXT(5,' PRSTDAT: ERROR WRITING FILE: ')
C IERR=1
C CALL CLOSE(LN)
C RETURN
C CALL TEXT(5,' PRSTDAT:IRCMX EXCEEDS MAXBUF. IRCMX = ',IRCMX)
C CALL TEXT(5,' CURRENT VALUE OF MAXBUF = ',MAXBUF)
C CALL TEXT(5,' IF YOU INCREASE MAXBUF,MAKE SURE YOU INCREASE THE')
C CALL TEXT(5,' DIMENSION OF BUF ACCORDINGLY.')
C STOP
C END
=====

```

```

0001 C=====
0002 SUBROUTINE HET(Z,ANMIN,NSAMP,FRQ,SSR)
0003 IMPLICIT COMPLEX (Z)
0004 DIMENSION Z(1)
0005 C
0006 C THIS SUBROUTINE PERFORMS COMPLEX HETEROODYNING ON THE
0007 C INPUT COMPLEX SERIES CONTAINED IN THE ARRAY Z.
0008 C
0009 DATA TWOPI/6.283185307/
0010 IF (FRQ.EQ.0.) RETURN
0011 PHIO=SIGN(1.,FRQ)*AMOD(ABS(FRQ)/SSR,1.)
0012 ZW=CEXF(CMPLX(0.,TWOPI*PHIO))
0013 PHIS=SIGN(1.,ANMIN*PHIO)*AMOD(ABS(ANMIN*PHIO),1.)
0014 ZS=CEXF(CMPLX(0.,TWOPI*PHIS))
0015 DO 10 N=1,NSAMP
0016 Z(N)=Z(N)*ZS
0017 ZS=ZS*ZW
0018 CONTINUE
0019 RETURN
0020 END
  
```

COMMAND QUALIFIERS

```

FORTRAN /NOOBJ/LIS/SHOW=(INCLUDE,NOMAF) TRDRIV1,TRUTILS1
/CHECK=(NOBOUNDS,OVERFLOW,NOUNDERFLOW)
/DEBUG=(NOSYMBOLS,TRACEBACK)
/STANDARD=(NOSYNTAX,NOSOURCE_FORM)
/SHOW=(NOFREPROCESSOR,INCLUDE,NOMAF)
/F77 /NOG_FLOATING /I4 /OPTIMIZE /WARNINGS /MOD_LINES /NOCROSS_REFERENCE /NOMACHINE_CODE /CONTINUATIONS=19
  
```

COMPIATION STATISTICS

```

Run Time: 18.35 seconds
Elapsed Time: 39.47 seconds
Page Faults: 23
Dynamic Memory: 164 pages
  
```

```
0001 SUBROUTINE QAR(LUN, ITEXT, ANS)
0002   BYTE ITEXT(1), IZER
0003   DATA IZER/0/
0004   DO 10 I=1,60
0005     IF (ITEXT(I).EQ.0) GO TO 20
0006     CONTINUE
0007     WRITE(LUN,30) (ITEXT(I1), I1=1, I-1), (IZER, I1=1, 60)
0008     FORMAT(60A1,$)
0009     READ(LUN,40,ERR=5) ANS
0010     FORMAT(015.0)
0011     RETURN
0012   END
```

28-May-1985 17:41:05 VAX-11 FORTRAN V3.2-37
5-Jul-1984 14:56:40 DISK\$USER1:CELDKIDG10AN.FOR:14

```

0001 SUBROUTINE QAX(LUN, ITEXT,ANS)
0002   BYTE ITEXT(1), IZER
0003   DATA IZER/0/
0004   DO 10 I=1,60
0005     IF (ITEXT(I).EQ.0) GO TO 20
0006   CONTINUE
0007   WRITE(LUN,30) (ITEXT(I1), I1=1, I-1), ( IZER, I1=1, 60)
0008   FORMAT(60A1,$)
0009   READ(LUN,40,ERR=5)ANS
0010   FORMAT(Z6)
0011   RETURN
0012   END

```

5 10 20 30 40

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5-Jul-1984 14:56:40

VAX-11 FORTRAN V3.2-37
DISK\$USER11:ELDRIDGE.DAN.FOR:14

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```
0001 SUBROUTINE QAI(LUN,ITEXT,IAN5)
0002   BYTE ITEXT(1), IZER
0003   DATA IZER/0/
0004   DO 10 I=1,60
0005     IF (ITEXT(I).EQ.0) GO TO 20
0006     CONTINUE
0007   WRITE(LUN,30) (ITEXT(I1), I1=1, I-1), (IZER, I1=1, 60)
0008   FORMAT(60A1,$)
0009   READ(LUN,40,ERR=5) IAN5
0010   FORMAT(I15)
0011   RETURN
0012   END
```

```

0001 SUBROUTINE QAA(LUN, ITEXT, IANS)
0002   BYTE ITEXT(1), IZER, IANS(1), IBUF(40)
0003   DATA IZER/0/
0004   DO 10 I=1, 60
0005     IF (ITEXT(I).EQ.0) GO TO 20
0006   CONTINUE
0007   WRITE(LUN, 30) (ITEXT(I), I=1, I-1), (IZER, I=1, 60)
0008   FORMAT(40A1, $)
0009   READ(LUN, 40, ERR=5) NCHR, (IBUF(NC), NC=1, NCHR)
0010   FORMAT(0, 40A1)
0011   IANS(NCHR+1)=0
0012   DO 50 I=1, NCHR
0013     IANS(I)=IBUF(I)
0014   CONTINUE
0015   RETURN
0016   END

```

28-May-1985 17:41:05 VAX-11 FORTRAN V3.2-37
5-Jul-1984 14:56:40 DISK\$USER1:[ELDRIDGE]QAN.FOR;14

```
0001 SUBROUTINE TEXIR(LUN,ITEXT,ANS)
0002   BYTE ITEXT(1),IZER
0003   DATA IZER/0/
0004   DO 10 I=1,60
0005     IF(ITEXT(I).EQ.0)GO TO 20
0006   CONTINUE
0007   WRITE(LUN,30)(ITEXT(I1),I1=1,I-1),(IZER,I1=I,60),ANS
0008   FORMAT(60A1,G12.4)
0009   RETURN
0010   END
0011
```



```
0001 SUBROUTINE TEXTX(LUN, ITEXT,ANS)  
0002   BYTE ITEXT(1), IZER  
0003   DATA IZER/0/  
0004   DO 10 I=1,60  
0005     IF (ITEXT(I).EQ.0)GO TO 20  
0006     CONTINUE  
0007     WRITE(LUN,30) (ITEXT(I1), I1=1, I-1), ( IZER, I1=I, 60),ANS  
0008     FORMAT(60A1,Z8.8)  
0009     RETURN  
0010   END
```

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```

0001 SUBROUTINE TEXT(LUN,ITEXT)
0002 BYTE ITEXT(1),IZER
0003 DATA IZER/0/
0004
0005 5 DO 10 I=1,72
0006 IF (ITEXT(I).EQ.0)GO TO 20
0007 10 CONTINUE
0008 20 WRITE(LUN,30)(ITEXT(I1),I1=1,I-1),(IZER,I1=1,72)
0009 30 FORMAT(72A1)
0010 RETURN
0011 END

```

0001	SUBROUTINE TEXTA(LUN,ITEXT,IANS)
0002	BYTE ITEXT(1),IZER,IANS(20)
0003	DATA IZER/0/
0004	DO 10 I=1,52
0005	IF(ITEXT(I).EQ.0)GO TO 20
0006	CONTINUE
0007	WRITE(LUN,30)(ITEXT(I1),I1=1,I-1),(IZER,I1=1,52),IANS
0008	FORMAT(72A1)
0009	RETURN
0010	END

5	10
	20
	30

0002	5	BYTE ITEXT(1), IZER
0003		DATA IZER/0/
0004		DO 10 I=1, 60
0005		IF (ITEXT(1).EQ.0) GO TO 20
0006	10	CONTINUE
0007	20	WRITE(LUN, 30) (ITEXT(I), I=1, I-1), (IZER, I=1, 60), IANS
0008	30	FORMAT(60A1, I6)
0009		RETURN
0010		END

END

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